Supplementary Tables

act lla	adil Ila	Donlingto	Linear regr	regression	M-W younger vs older fathers	s older fathers	M-W younger fathers vs simulations	s vs simulations	M-W older fathers vs simulations	s vs simulations
adó lao		ויפטווסמופ	р	β	d	β	d	β	Ф	В
	6 lines	·	0,0022	0,158	1,3E-04	-0,070	4,9E-04	-0,053	0,6839	0,004
Lymphoblastoid cells	C0202	~	0,0305	0,110	1,7E-03	-0,042	2,7E-03	-0,033	0,83	0,002
	C0202	2	0,0285	0,111	9,9E-04	-0,050	2,8E-04	-0,046	0,70	-0,004
Neural precursor	BG01	1	0,0265	0,108	1,5E-03	-0,057	9,5E-04	-0,049	0,84	-0,002
cell	BG01	7	0,0257	0,111	7,5E-04	-0,051	7,6E-04	-0,042	0,93	-0,001
	BG01		7,7E-04	0,171	3,6E-04	-0,057	5,0E-05	-0,055	0,37	600'0-
	BG02	~	0,0036	0,148	5,2E-04	-0,048	4,4E-05	-0,047	0,27	600'0-
Embryonic stem cell	BG02	2	0,0047	0,146	4,7E-03	-0,044	1,1E-03	-0,042	0,35	600'0-
	H7	ı	0,0036	0,146	3,2E-03	-0,051	1,5E-04	-0,055	0,16	-0,015
	6Н	•	0,0021	0,155	7,1E-04	-0,050	1,3E-05	-0,054	0,15	-0,013
	iPS4	1	6,1E-04	0,172	1,1E-03	-0,054	5,6E-05	-0,056	0,31	-0,011
Induced	iPS4	2	0,0046	0,143	3,0E-04	-0,065	8,4E-05	-0,059	0,68	-0,004
cell cell	iPS5	~	8,9E-04	0,167	3,6E-04	-0,060	3,2E-04	-0,051	0,93	-0,001
	iPS5	2	8,7E-04	0,171	5,5E-03	-0,045	2,7E-03	-0,040	0,65	-0,004

Supplementary Table 1. Paternal age effect on de novo mutation replication timing measured in 6 cell types

cell lines. The linear regression column contains p-values and estimates (β) for the parsimonious model described in Methods. The The source of the replication timing data was Koren et al. 1 for the 6 lymphoblastoid cell lines (1st row) and Ryba et al. 2 for all other M-W test columns contain the p-values and estimated difference using a Mann-Whitney test between the distribution of mutation replication timing values of: offspring of younger (<28 years old) vs older (≥28 years old); offspring of younger fathers vs simulations; offspring of older fathers vs simulations. Significant p-values are highlighted in bold font

	$P(\boldsymbol{\beta}_{r,t} \neq 0)$
T→G	0.54
C→T	0.00696
G→C	0.0502
T→C	0.221
T→A	0.025
C→A	6.06 x 10 ⁻⁸
CpG→TpG	1.98 x 10 ⁻⁸

Supplementary Table 2. Predictive power of local substitution rates.

Predictive power of primate substitution rates for local *de novo* mutation rates using the Poisson regression model described in the Supplementary Note. Only $S \rightarrow W$ and $W \rightarrow W$ substitutions have significant predictive power for local *de novo* mutation rates.

	$\widehat{oldsymbol{eta}_{ ho,oldsymbol{t}}}$	$P(\widehat{\beta_{\rho,t}}\neq 0)$	\widehat{f}_t
G→A	-2.485 x10 ⁻⁵	0.23	1.034
T→A	2.597 x 10 ⁻⁶	0.649	0.891
C→A	-9.857 x 10 ⁻⁵	< 2 x 10 ⁻¹⁶	1.014
CpG→CpA/TpG	-0.00449	< 2 x 10 ⁻¹⁶	0.988

Supplementary Table 3. Dependency of local mutation rates on recombination rates

Summarizes the estimated dependency of local mutation rates on recombination rates. Only $C \rightarrow A$ and $CpG \rightarrow TpG$ exhibit a significant dependency.

Log Likelihood	GoNL uniform	Primate $r_{\scriptscriptstyle t,i}$		Corrected $\mu_{t,i}$ (%/%)*	
Recomb. rates	No	No	Male	Sex-Averaged	Female
$\log L_{T>G}$	-1,591.0	-1,605.3	-1,584.6 (0.4% / 1.3%)	-1,584.6 (0.4% / 1.3%)	-1,584.6 (0.4% / 1.3%)
$\log L_{G>A}$	-2,974.5	-2,961.7	-2,959.2 (0.5% / 0.1%)	-2,959.2 (0.5% / 0.1%)	-2,957.7 (0.6% / 0.1%)
$\log L_{G>C}$	-1,735.6	-1,736.5	-1,728.3 (0.4% / 0.5%)	-1,728.4 (0.4% / 0.5%)	-1,728.4 (0.4% / 0.5%)
$\log L_{T>C}$	-3,176.2	-3,183.5	-3,152.0 (0.8% / 1.0%)	-3,152.0 (0.8% / 1.0%)	-3,152.0 (0.8% / 1.0%)
$\log L_{T>A}$	-1,439.8	-1,437.7	-1,434.5 (0.4% / 0.2%)	-1,434.5 (0.4% / 0.2%)	-1,434.5 (0.4% / 0.2%)
$\log L_{C>A}$	-1,852.2	-1,836.7	-1,832.5 (1.1% / 0.2%)	-1,831.9 (1.1% / 0.3%)	-1,831.7 (1.1% / 0.3%)
$\log L_{CpG>CpA/TpG}$	-2,696.2	-2,439.6	-2,435.8 (9.7% / 0.2%)	-2,435.0 (9.7% / 0.2%)	-2,434.9 (9.7% / 0.2%)
Total	-15,465.6	-15,201.0	-15,126.9 (2.2% / 0.5%)	-15,125.6 (2.2% / 0.5%)	-15,123.8 (2.2% / 0.5%)

^{*} In parenthesis is the percent change of log likelihood of $\mu_{t,i}$ compared to GoNL uniform model and uncorrected primate rate model $r_{t,i}$.

Supplementary Table 4. Likelihood of the observed data under different mutation rate models

Likelihood of the observed de novo mutation data by substitution type based on (a) a uniform mutation rate model derived from the observed mutations, (b) the uncorrected primate rate matrix $r_{t,i}$ and (c) the computed mutation rate matrix $\mu_{t,i}$.

Supplementary Note

Mutation rate map

For each 1Mb window i, a substitution rate matrix was inferred using the context-dependent primate substitution model described in Duret *et al.* ³for seven types of substitutions, parameterized by $r_{t,i}$ with t in {T \rightarrow G, G \rightarrow C, T \rightarrow C, T \rightarrow A, C \rightarrow A, C \rightarrow T, CpG \rightarrow TpG (to account for hyper-mutability of CpG sites)}.

First, we tested if the observed *de novo* mutation rates co-vary with primate substitution rates across the genome using the following Poisson regression model with log link function:

$$\begin{split} \log(n_{t,i}) \\ &= \beta_{r,t} r_{t,i} + \beta_{r,t,0} + \log(N_{t,i}) \quad \text{for } t \neq \mathsf{CpG} \rightarrow \mathsf{TpG} \\ &= \beta_{r,t} \big(r_{t,i} + r_{C \rightarrow T,i} \big) + \beta_{r,t,0} + \log(N_{t,i}) \quad \text{for } t = \mathsf{CpG} \rightarrow \mathsf{TpG} \end{split}$$

where $n_{t,i}$ is the observed count of de novo mutations of type t in window i, $r_{t,i}$ is the substitution rate of type t in window i, and $N_{t,i}$ is the number of sites at which de novo mutations of type t can be detected with high confidence in window i. The offset term $\log{(N_{t,i})}$ was added since the number of called de novo mutations is dependent on detection power. The C \rightarrow T mutation of CpG sites requires special treatment since it can be attributed to context-independent C \rightarrow T substitution as well as hyper-mutability of CpG.

The primate substitution rates in the above Poisson regression model only had significant predictive power for local *de novo* mutation rates for S \rightarrow W and W \rightarrow W substitutions (Supplementary Table 2). For this reason, we only estimated local mutation rates based on the primate substitution rate for substitutions in $t_{SW} = \{T\rightarrow A, C\rightarrow A, C\rightarrow T, CpG\rightarrow TpG\}$. For the rest of substitutions (t in $\{T\rightarrow G, T\rightarrow C, G\rightarrow C\}$), we used the genome-wide averaged mutation rates r'_t estimated from our observed mutations:

$$r'_{t} = \frac{\sum_{i} n_{t,i}}{\sum_{i} N_{t,i}} \cdot \frac{1}{c}$$

$$c = \frac{\sum_{i} \sum_{t \in t_{SW}} n_{t,i}}{\sum_{i} \sum_{t \in t_{SW}} r_{t,i} N_{t,i} + \sum_{i} r_{C \to T,i} N_{CpG \to TpG,i}}$$

where *c* is a scaling factor to convert between *de novo* mutation rates and instantaneous substitution rates.

Second, we corrected for the biases due to local recombination rates. The observed local *de novo* mutation rates were not significantly correlated with recombination rates when considering each type of substitutions separately (Bonferroni-corrected p-value > 0.05). However, local substitution rates $r_{t,i}$ depend significantly on local sex-averaged recombination rates ρ_i for $t = C \rightarrow A$ and CpG \rightarrow TpG (Supplementary Table 3). To eliminate the dependency on recombination rate, we fit the following linear regression model:

$$r_{t,i} = \beta_{\rho,t} \, \rho_i + \beta_{0,t}$$

and residualized $r_{t,i}$ by subtracting the ρ_i -dependent term.

The final formula we used to compute the mutation rates for each 1Mb window *i* is then:

$$\mu_{t,i} = (r_{t,i} - \beta_{\rho,t} \rho_i) \cdot f_t \text{ for } t \text{ in } \{C \rightarrow A, CpG \rightarrow TpG\}$$

$$\mu_{t,i} = r_{t,i} \cdot f_t$$
 for t in $\{T \rightarrow A, C \rightarrow T\}$

$$\mu_{t,i} = r'_t \text{ for } t \text{ in } \{T \rightarrow G, T \rightarrow C, G \rightarrow C\}$$

where f_t is a global scaling factor for substitution of type t to match the observed frequencies of different types of *de novo* mutations (Supplementary Table 4). In particular, A \rightarrow T mutation is over-represented in primate substitutions by 12% compared to our *de novo* data. For each t in t_{SW} , f_t is defined to satisfy the following conditions:

$$\begin{split} & \sum_{i} \mu_{t,i} N_{t,i} = \frac{1}{c} \sum_{i} n_{t,i} \quad \text{for } t \text{ in } \{\mathsf{T} \rightarrow \mathsf{A}, \ \mathsf{C} \rightarrow \mathsf{T}, \ \mathsf{C} \rightarrow \mathsf{A}\} \\ & \sum_{i} \left(\mu_{t,i} + \mu_{C \rightarrow T,i} \right) N_{t,i} = \frac{1}{c} \sum_{i} n_{t,i} \quad \text{for } t = \mathsf{CpG} \rightarrow \mathsf{TpG} \end{split}$$

Finally, the mutation rate μ was scaled so that the overall mutation rate across the autosome is 1.2×10^{-8} per nucleotide per generation.

To evaluate the fit of the estimated mutation rates to observed *de novo* mutations, we examined the likelihood L_t of the observed data given mutation rates, assuming homogenous Poisson process for each type of mutation t within each window i:

$$\begin{split} L_t \big(data \big| \mu_{t,i} \big) \\ &= \prod_i Poisson \left(\left. n_{t,i} \right| \lambda = \frac{1}{c'} \mu_{t,i} N_{t,i} \right) \quad \text{for } t \neq \mathsf{CpG} \rightarrow \mathsf{TpG} \\ &= \prod_i Poisson \left(\left. n_{t,i} \right| \lambda = \frac{1}{c'} \left(\mu_{t,i} + \mu_{C \rightarrow T,i} \right) N_{t,i} \right) \quad \text{for } t = \mathsf{CpG} \rightarrow \mathsf{TpG} \\ c' &= \frac{\sum_i \sum_t \mu_{t,i} N_{t,i} + \sum_i \mu_{C \rightarrow T,i} N_{CDC \rightarrow TDC,i}}{\sum_i \sum_t \mu_{t,i} N_{t,i} + \sum_i \mu_{C \rightarrow T,i} N_{CDC \rightarrow TDC,i}} \end{split}$$

The likelihood of the observed data under different models is summarized in Supplementary Table 4.

We estimated functional mutation rates in protein-coding region for autosomal protein-coding transcripts (downloaded from Ensembl 4 v74). Excluding 24,508 transcripts (3,808 genes) outside our analysis windows for bias correction, we computed bias-corrected mutations rates for a total of 54,310 transcripts (15,462 genes). For maximum coverage of genes, however, we provide two additional functional mutation rates based on uncorrected local primate substitution rates $r_{t,i}$ and the uniform genome-wide averaged mutation rates r_t derived from our observed data.

For each transcript, the local mutation rate was determined by the 1Mb genomic window that overlapped the coordinate of midpoint between transcription start and end sites. Based on this rate, all possible nonsense, missense, synonymous and 4-fold degenerate synonymous mutations were examined with respect to the reference genome, and their mutation rates were aggregated over the entire transcript.

While we assumed the equal rate of $\mu_{A\to G,i}$ and complementary $\mu_{T\to C,i}$ in non-coding region, we adjusted for their strand bias in protein-coding region as follows:

$$\begin{split} \mu_{A \to G, i}^{tx} &= \frac{N_A^{tx} + N_T^{tx}}{N_A^{tx}} \frac{\gamma_{sb}}{1 + \gamma_{sb}} \mu_{T \to C, i}^{nc} \\ \mu_{T \to C, i}^{tx} &= \frac{N_A^{tx} + N_T^{tx}}{N_A^{tx}} \frac{1}{1 + \gamma_{sb}} \mu_{T \to C, i}^{nc} \end{split}$$

$$\gamma_{sb} = \frac{n_{A \to G}^{tx}}{n_{T \to C}^{tx}}$$

where $\mu_{T \to C,i}^{nc}$ (= $\mu_{A \to G,i}^{nc}$) is the local mutation rate of A:T \to G:C in non-coding region, $\mu_{T \to C,i}^{tx}$ and $\mu_{A \to G,i}^{tx}$ are the local mutation rates of T \to C and A \to G in protein-coding with respect to the transcribed strand, N_A^{tx} and N_T^{tx} are the total numbers of protein-coding A and T bases in transcribed strand across the autosomes, and $n_{T \to C}^{tx}$ and $n_{A \to G}^{tx}$ are the genome-wide counts of observed T \to C and A \to G de novo mutations with respect to the transcribed strand in our dataset. γ_{sb} was estimated to be 1.389 and $N_A^{tx}/(N_A^{tx}+N_T^{tx})$ to be 0.543 in our data.

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